

Domain Parking: Not as Malicious as Expected

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Executive Summary

Domain parking is the practice of assigning a nonsense address to a domain when it is not in use in order to keep it ready for "live" use. This practice is peculiar because it indicates someone has administrative control over the domain name, does not have hardware ready to respond to requests, but wants the domain to appear active. A more appropriate response would seem to us to be that the domain does not exist. This mismatch between expected benign behavior (no such domain) and actual observed behavior (parking) made us suspicious. In this paper we discuss scalable detection methods for domain names parking on reserved IP address space, and then using this data set evaluate whether this behavior appears to be indicative of malicious behavior.

We find that during the month of January 2014 only 21,328 unique domains exhibited parking on reserved address space, out of over 610 million total unique observed domains. Thus, parking appears to be an uncommon Internet behavior with only 0.0035% of domains exhibiting parking on reserved IP addresses. Of these 21,328 domains, relatively few were observed listed on any of 16 domain black lists any time from January 1 to February 28, 2014. Only 1,563, or 7.3%, were listed in this time period. Therefore, we conclude that parking is a poor indicator of malicious activity, or at least not an indicator of any kind of malicious activity usually examined by any public list of malicious domain behavior.

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1 Introduction

When a domain is "parked" on an IP address, the IP address to which the domain resolves is inactive or otherwise not owned by the domain owner. This is a common practice when a user first registers a domain – the registrar does not know what IP to supply as an answer, but supplying some answer prevents errors.

The domain name system permits a variety of different mechanisms which help provide resiliency to distributed architectures. Often these have legitimate uses, but malicious actors are equally able to adopt successful techniques. Usually the malicious use case is sufficiently different that the type of use can be teased apart. Suspicious domain name parking is no different; herein we present a method for finding it in historical passive DNS data.

Malicious actors seem to have adopted this technique for similar error suppression goals as the benign use case. Although it is suppression of different errors, such as evading detection before the number of infected machines reaches the desired number or while the command and control structure is not yet in place. We present a method for detection of domains that exhibit parking and a mechanism for distinguishing legitimate from suspicious use.

This parking destination, reserved IP space, is quite different from parking a domain on someone else's IP space. To our knowledge, there has been one study on parking domains for illicit ad revenue, which appears to happen on a large scale of 4 million domains [1]. However, from the authors' description this appears to be more like typosquatting (as described in Szurdi et. al. [2]) than resolution error suppression, as the authors describe the "dark side of domain parking" as monetized "whenever web users type in those domain names (probably accidentally) in the browser's address bar, the parking service resolves the domains to advertisement laden pages" [1, p. 1]. We are not aware of other studies of domain parking, except that some fast-flux identification algorithm studies cited domain parking as an obstacle [3, 4].

Parking on reserved IP space is sufficiently uncommon that it is somewhat difficult to find, at only 0.0035% of unique domains observed. This difficulty is not so much because it is infrequent but that the IP addresses commonly used for parking, such as the 127.0.0.0/8 block or those reserved in RFC 1918 [5] are also used for several other more common uses of the DNS, such as delivering real-time DNS black list results [6]. This introduces noise into any detection technique since it is not so simple as just finding domains that pointed to reserved address space at some time and then changed.

CIDR block	Justification
10.0.0.0/8	RFC 1918 [5]
127.0.0.0/8	RFC 1700 [9]
169.254.64.0/18	RFC 3927 [10]
172.16.0.0/12	RFC 1918 [5]
192.168.0.0/16	RFC 1918 [5]

Table 1: Private IP address space

2 Method

The main prerequisite for our method is a large source of passive DNS trace data. In order to calculate over large data volumes, we take several simplifying steps. Data is ingest in nmsgtool format [7], including source DNS server and precise time range the response was valid, at a rate of about 35 GB per day. Unique resource record sets (RRsets) are extracted from the DNS messages and extraneous fields are removed, leaving just the fields for rname, TTL, type, and rdata [8]. A list of unique RRsets per day based on these fields is approximately 2 GB in our data source.

Then, we load the RRsets with type of A (IPv4 answer) for January 2014 into a PostgreSQL database. The table has fields for the four RRset fields as well as day observed. Since RRsets are unique per day, if an identical RRset was observed on multiple days it will appear in the database for each day observed. This structure permits a course-grained time series view with enough data to detect patterns but enough summarization that calculation is practical.

In order to start our search for parking on private IP address space, we query the database for all RRsets where the rdata is in the IP set indicated in Table 1. Most of the results are not actually parking. Answers in private IP space are used to encode various kinds of non-location data, such as responses to lookups on DNSBLs, and for other administrative reasons in content distribution networks and hosting companies. We created a list by expert human analysis to remove these irrelevant domains from the results. Table 2 lists the second-level domains (SLDs) that were removed.

The process so far yields a list of RRsets with rdata in private IP space and rname domains that do not have a known use. We search for all other RRsets with the same domains in the rname field. Any results will have publicly routeable IP addresses, and thus at some point in the month have transitioned between private and routable IP address space. We consider these domains to have exhibited parking behavior on private IP address space.

abuseat.org	httpbl.org	schpider.com
ahbl.org	invaluement.com	senderscore.com
anubisnetworks.com	isipp.com	sonicwall.com
apews.org	ja.net	sophosxl.com
barracudacentral.org	jtripper.net	sorbs.net
bl.rptn.ca	junkemailfilter.com	spamcop.net
blocklist.de	kaspersky-labs.com	spameatingmonkey.net
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borderware.com	lsu.edu	spamhaus.org
ciphertrust.net	mail-abuse.com	spamrats.com
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drweb.com	njabl.org	ufl.edu
dsadns.net	nszones.com	uribl.com
dscwl.net	pacanka.com	validatorsearch.verisignlabs.com
dsintll.net	qualcomm.com	vircom.com
dsl.cantv.net	quorum.to	webcfs00.com
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Table 2: Domains that were removed from analysis

For each domain name that has exhibited parking behavior, we can generate a course-grained time series of the behavior to categorize what occurred. Table 3 demonstrates some sample behavioral groupings. P indicates a day where the only rdata was in private IP address space, G indicates a day where the only rdata was in globally routeable IP address space, and X indicates a day where both address types were observed, indicating a day a change between parking and active occurred.

Analysis of the domains found to exhibit parking mostly included simple text

January:	1-8	9-16	17-24	25-31
Activation on Jan 19	PPPPPPPP	PPPPPPPP	PPXGGGGG	GGGGGGG
Deactivation on Jan 19	GGGGGGG	GGGGGGG	GGXPPPPP	PPPPPPP
com.alextringham	GGGGGGG	GGGGGGG	GGGGXPPX	PXPXPPP
cn.proxyie	GGXXXXXX	GGGXGXGG	GGPGGXGX	XGGGXGX
net.homeip.bnlv	GGGGGPGG	GGGGGGG	PGPPGGGG	GGGGPGG

Table 3: Example parking behavior patterns, January 2014. G := only globally routable IPs observed for a domain on a given day. P := only privately reserved IPs observed. X := both observed on same day.

matching on lists of malicious domains. While we have expressed our doubts about the soundness of evaluating an approach by comparing it to black lists [11], we have mitigated this analysis error by including as many lists as possible and limiting our assumptions of the information provided by this comparison.

Analysis of routable IP addresses includes geolocation and ASN attribution information. Geolocation is derived from the public MaxMind GeoLite2 free geolocation data from January 28, 2014 [12]. ASN attribution is derived from our publicly available IP-to-ASN mapping published for January 31, 2014, itself derived from the RouteViews [13] and RIPE NCC RIS [14] data. The baseline mapping of ASNs across all IP space uses our open-source SiLK [15] tools for prefix maps and IP sets [16].

3 Results

We applied our method to all unique domains observed in our passive DNS data source for the month of January 2014. This data set contains 610 million total unique domains. After applying our method described above, 21,328 unique domains exhibit parking, or 0.0035% of the total unique domains. This number includes domains that should not publicly resolve, such as .local, but which did in fact have both private and public DNS answers during the period of observation.

An additional 34 domains were found to appear to exhibit parking behavior, however all 34 domains were extremely popular domains listed in the Alexa top 100 at the time [17]. We did not count these popular domains in the 21,328 that we considered to exhibit parking behavior.

In order for some assessment of known maliciousness, we checked these domains that exhibited IP address parking on private address space against 16 domain-based lists of malicious domains. 1,563 domains appeared on at least one such list

¹http://routeviews-mirror.cert.org/pmap/2014/01/20140131.bgp.pmap

Count	% of Parking	% of All Domains
8594	40.2831%	65.7351%
2651	12.4262%	20.7651%
1045	4.89828%	1.9186%
842	3.94675%	0.2514%
662	3.10303%	0.1268%
660	3.09365%	0.0430%
463	2.17024%	0.5156%
441	2.06712%	0.0931%
336	1.57495%	0.2265%
282	1.32183%	0.2541%
	8594 2651 1045 842 662 660 463 441 336	8594 40.2831% 2651 12.4262% 1045 4.89828% 842 3.94675% 662 3.10303% 660 3.09365% 463 2.17024% 441 2.06712% 336 1.57495%

Table 4: Top 10 TLDs by number of domains exhibiting IP-address parking on private address space

between January 1 and February 28, 2014. We allowed some additional time beyond when the domains exhibited parking in order to allow a better chance the domain would be discovered by a list, as there is some expected lag time for detection.

In order to assess some features of the network connectivity and domain structure, the 21,328 domains can be broken down by top-level domain (TLD) and whether the domain is hosted by a known dynamic DNS provider. Table 4 details the breakdown of the parking domains by TLD. We compared the 21,328 domains to a list of 71 known dynamic DNS providers as well: 353 domains were hosted in this way. The bulk hosted on two providers: 111 on dyndns.org and 191 on some name affiliated with no-ip. These are the two biggest dynamic DNS providers generally.

We can also characterize the IP addresses used to host the domains while they were routable. 41,170 unique public IP addresses were used as the routable IP addresses for some domain that exhibited parking (on private IP addresses). Each IP address had an average of 1.38 domains pointing to it, though there is clearly a heavily skewed distribution, as displayed in Table 5. We can also characterize these IP addresses by their geographic location, as best as we can determine it. The IP addresses were distributed across 164 countries, also in a long-tail distribution. Table 6 displays the 10 most common locations.

The autonomous system number (ASN) of the public IP addresses used, ASNs that announced the IP addresses were examined with the top 10 in Table 7. While the ASN counts are more evenly distributed, there is a bias of some kind towards certain ASNs. The selection of destination IP addresses is not distributed randomly across ASNs, some networks host many times the proportion of these locations

# of domains	# of IPs with X domains
X=1	36765
$X \le 10$	4169
$X \le 50$	188
$X \le 100$	20
X > 100	28

Table 5: Distribution of domains per IP address

Country Code	# of IP Addresses
US	17438
RU	3152
UA	2163
CN	1508
DE	1273
BR	907
CA	865
GB	809
TW	795
NL	734

Table 6: Top 10 countries in which IP addresses of domains exhibiting parking were hosted, as geolocated on Jan 28, 2014

than is explainable purely by chance.

4 Conclusions

The number of domains exhibiting parking on private IP addresses is quite small. And although the behavior appears to be distributed in ASNs and locations non-randomly, it does not appear to be a consistent indicator of malicious activity. The process for finding domains genuinely exhibiting parking is somewhat tedious, with a fair amount of manual review and whitelisting of domains for non-location uses that confuse the results. The process also requires a relatively long observation window, as the observation must allow enough time for the domain to change rdata. These two features impose a relatively high cost on finding parking domains, while there are not clear benefits to discovering them. The domains do not have a clear malicious intent, there are not many of them, and the domains are general uninteresting by our *prima facie* expert analysis. This particular kind of parking

ASN	Count	% of parking IPs	% of Internet assigned to ASN
AS6079	1574	3.82317%	0.02171%
Unknown	881	2.13991%	37.63242%
AS6517	834	2.02575%	0.00833%
AS22773	799	1.94073%	0.27731%
AS5739	732	1.77799%	0.00305%
AS8075	629	1.52781%	0.03512%
AS4134	601	1.45980%	2.52874%
AS15003	585	1.42094%	0.04291%
AS3462	525	1.27520%	0.28541%
AS46606	519	1.26063%	0.01507%

Table 7: Top 10 ASNs announcing routable IP addresses used by domains that exhibit parking. ASN mappings are as of January 15, 2014.

behavior does not appear to be useful to detect. The malicious behavior detected in this way would very likely be easier to detect by existing methods.

It is possible that the domains exhibiting this kind of parking are actually malicious, but simply are not found by any other method that would have them end up on the black lists we compare against. As lists of malicious behavior are mostly idiosyncratic [11], this is not entirely unlikely. We have made the complete list of domains available² in case another analysis can determine if they are, in fact, interesting. If so, we would welcome being proven wrong about their uninterestingness.

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